AUEB M.Sc. in Data Science (part-time)

Semester: Summer 2020

Course: Big Data Systems and Techniques

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General comments

Copying files from / to remote Google Cloud VMs

Using the *Google Cloud SDK* (requires installation on your platform of choice), one can copy files from remote instances in the following fashion:

```
WINDOWS shell
gcloud compute scp "root@s01:<remote_file_path>" "<local_file_path>"
```

Task 1 (5% - 15% if you use Google Cloud SQL) - Get the data

As deliverable describe the commands used to insert the data in postgres.

Using the local PostgreSQL instance

Getting the data

The data has been retrieved by the course Wiki page in the following fashion:

```
BASH shell

BASH shell

Wget --user <wiki_use> --password <wiki_pass>
https://bitbucket.org/dinosar/bigdatasystemscourse/downloads/products.sql.zip
```

Next, we unzip the *products.sql.zip* archive:

```
BASH shell
unzip products.sql.zip
```

Performing PostgreSQL restore operations

Accessing the PostgreSQL CLI

```
BASH shell sudo -u postgres psql
```

Creating the database user

```
PSQL shell

CREATE USER dinosar WITH PASSWORD 'forsharingpurposes';
```

Creating database "products":

```
PSQL shell
CREATE DATABASE products;
```

Restoring the "products.sql" dump file:

```
BASH shell
sudo -u postgres psql products < products.sql
```

Script output:

```
SET
SET
SET
SET
SET
SET
CREATE EXTENSION
COMMENT
```

```
SET
SET
SET
CREATE TABLE
ALTER TABLE
COPY 2080734
REVOKE
REVOKE
GRANT
```

Verification

bdsat-homework-report.md

List databases from within the PostgreSQL shell:

```
PSQL shell
```

Script output:

Switch to database "products":

```
PSQL shell
\c products
```

Describe schema tables:

```
PSQL shell
```

Script output:

Verify number of rows in table "temp_products" matches count:

```
PSQL shell

SELECT COUNT(*) FROM temp_products;
```

Script output:

```
count
-----
2080734
(1 row)
```

Quit PostgreSQL shell:

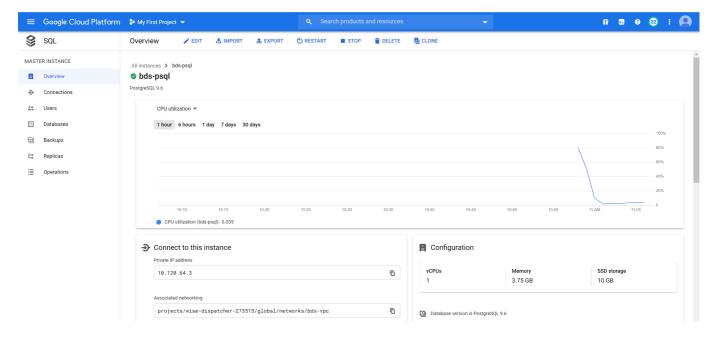
```
PSQL shell
```

Using Google Cloud SQL

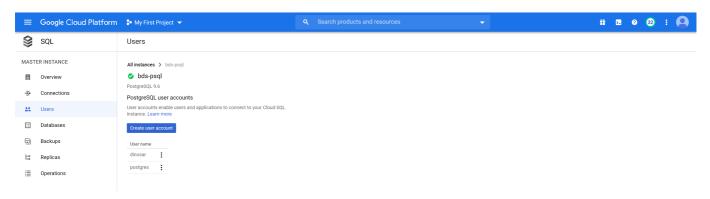
We created a Google Cloud SQL managed PostgreSQL instance by following these steps:

- In Google Cloud Console, we selected SQL -> Create instance.
- We selected *PostgreSQL* as the DB engine of choice.
- We supplied the instance characteristics:
 - o Name: bds-psql

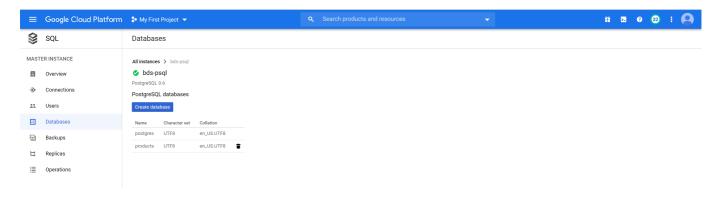
- o Version: 9.6
- Region: europe-west1-b
- Connectivity: private IP in bds-vpc
- Resources: 1 vCPU / 3.75 GB RAM / 10 GB SSD storage
- We ended up with the following instance:



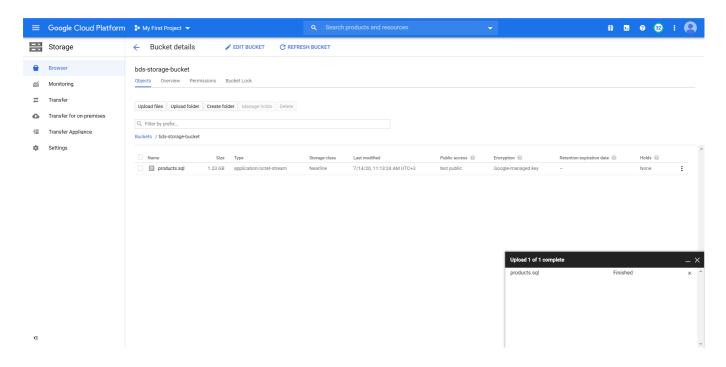
Next, we create the PostgreSQL user "dinosar":



• In the sequel, we created the database "products":



• In order to restore the "products.sql" dump, we had to upload it to a Google Cloud Storage bucket:



Finally, we restored the dump after we commented-out the following lines:

```
CREATE EXTENSION IF NOT EXISTS plpgsql WITH SCHEMA pg_catalog;
COMMENT ON EXTENSION plpgsql IS 'PL/pgSQL procedural language';
```

The reason is that, due to insufficient permissions, these two (2) lines of code prevented the dump from being restored.

Task 2 - Create a parquet file (10%)

Code

The code and process for this task can be found in *Jupyter Notebook Task-2.ipynb*.

Output

printSchema()

```
root
|-- product_id: integer (nullable = true)
|-- name: string (nullable = true)
|-- upc_id: string (nullable = true)
|-- descr: string (nullable = true)
|-- vendor_catalog_url: string (nullable = true)
|-- buy_url: string (nullable = true)
|-- manufacturer_name: string (nullable = true)
|-- sale_price: decimal(38,18) (nullable = true)
|-- retail_price: decimal(38,18) (nullable = true)
|-- manufacturer_part_no: string (nullable = true)
|-- country: string (nullable = true)
```

```
|-- vendor_id: integer (nullable = true)
|-- category_name: string (nullable = true)
|-- category_code: string (nullable = true)
|-- category_id: integer (nullable = true)
```

Output

HDFS Is operations

```
BASH shell

root@s01:~# hdfs dfs -ls output/
Found 1 items
drwxr-xr-x - root supergroup 0 2020-07-13 17:11
output/query_shoes.parquet
```

"query_shoes.parquet" is a directory, listing directory contents too:

Task 3 - ML (15% data processing, 25% algorithm training)

Code

The code and process for this task can be found in *Jupyter Notebook Task-3.ipynb*.

The task requires that we perform classification on features of the filtered dataset acquired from a previous step. Since we will be using the frozen model for *Task 5*, we shall use the textual description of each example as the feature set for our classification task.

Preprocessing

Following a NLP approach to our problem and using text tokens as features requires that we pre-process the text in a manner that reduces noise and normalizes the vocabulary that we will be using. To this end, we incorporate in our classification pipeline the following custom classes that inherit from *pyspark.ml.Transformer*:

• LowercaseTransformer: the pipeline stage that converts a string column to lowercase.

- RemoveHTMLTransformer: the pipeline stage that removes HTML from a string column.
- RemoveNonAlphanumericTransformer: the pipeline stage that removes punctuation and excessive whitespace characters from a string column.

The classes are packaged under src/Preprocessing/Tranformers.

A comment on the challenges faced during the implementation of the Transformers is that some external *Python* packages (e.g. *BeautifulSoup*, for *HTML* removal) had to be installed on the cluster environment in a "hacky" way (see notebook for details).

Testing our preprocessing pipeline, we end up with the following examples of shoes text descriptions (showing first 3 rows):

• Before applying preprocessing:

Double strap design in GG fabric and leather with signature web detail and cozy rubber sole.; Adjustable double grip-tape hook-and-loop strap closure; GG fabric and leather upper; Rubber sole; Padded insole; Made in Italy

Just because they're in the house doesn't mean their little feet stop moving! They'll be stylin' while keepin' it cozy in the Everest Pablo slipper. 100% boiled wool upper adorned with a colorful knit design. Slip-on design for easy on-and-off wear. Breathable wool lining helps keep feet dry and cool. Latex footbed with wool covering to keep little feet warm and comfy. Durable rubber outsole. Imported. Machine wash gentle, air dry. Measurements: ; Weight: 5 oz ; Circumference: 10 in ; Shaft: 5 in ; Product measurements were taken using size 33 (US 2 Little Kid), width M. Please note that measurements may vary by size.

A handsome lace-up oxford is crafted with a lightweight, flexible sole for play-ready appeal.
Leather upper/suede lining/rubber sole.
Kids' shoes.

• After applying preprocessing:

double strap design in gg fabric and leather with signature web detail and cozy rubber sole adjustable double grip tape hook and loop strap closure gg fabric and leather upper rubber sole padded insole made in italy

just because they re in the house doesn t mean their little feet stop moving they ll be stylin while keepin it cozy in the everest pablo slipper 100 boiled wool upper adorned with a colorful knit design slip on design for easy on and off wear breathable wool lining helps keep feet dry and cool latex footbed with wool covering to keep little feet warm and comfy durable rubber outsole imported machine wash gentle air dry measurements weight 5 oz circumference 10 in shaft 5 in product measurements were taken using size 33 us 2 little kid width m please note that measurements may vary by size

a handsome lace up oxford is crafted with a lightweight flexible sole for play

ready appeal leather upper suede lining rubber sole by jumping jacks imported kids shoes

We could go on with preprocessing (stopword removal, stemming / lemmatization etc.) but this is beyond the scope of this assignment.

Algorithm training

Dealing with a mult-class classification problem, we opted for using the following algorithms:

- A One-vs-Rest classifier with Logistic Regression
- A Random Forest classifier

For initial / manually training of the algorithms, we performed a 80 / 20 train / test dataset splitting and we acquired the following results:

- One-vs-Rest with Logistic Regression classifier.
 - Parameters:
 - Vectorizer *numFeatures*: 5000
 - Logistic Regression *maxIter*: 200
 - Logistic Regression *regParam*: 0.01
 - Metrics

Metric	Value
Accuracy	0.794
Error	0.206
Precision	0.758
Recall	0.706
F1-score	0.731
AUC	0.895

- Random Forest classifier:
 - Parameters:
 - Vectorizer *numFeatures*: 5000
 - Random Forest *numTrees*: 200
 - Metrics:

Metric Va	lue
-----------	-----

Metric	Value
Accuracy	0.474
Error	0.526
Precision	0.081
Recall	0.954
F1-score	0.149
AUC	0.716

Clearly, we shall continue with optimizing *One-vs-Rest* with *Logistic Regression*.

In the sequel, we implemented verbose *cross-validation* (code copied from the course Wiki) on our best model, acquiring the following output and results:

• Output:

```
Comparing models on fold 1
params: {'regParam': 0.1, 'numFeatures': 100}
                                                accuracy: 0.554878
                                                                    avg: 0.554878
params: {'regParam': 0.01, 'numFeatures': 100}
                                                accuracy: 0.574099
                                                                    avg: 0.574099
params: {'regParam': 0.1, 'numFeatures': 250}
                                                accuracy: 0.617287
                                                                    avg: 0.617287
params: {'regParam': 0.01, 'numFeatures': 250}
                                                accuracy: 0.643188
                                                                    avg: 0.643188
params: {'regParam': 0.1, 'numFeatures': 500}
                                                accuracy: 0.686200
                                                                   avg: 0.686200
params: {'regParam': 0.01, 'numFeatures': 500}
                                                accuracy: 0.713742
                                                                    avg: 0.713742
params: {'regParam': 0.1, 'numFeatures': 1000}
                                                accuracy: 0.730735
                                                                    avg: 0.730735
params: {'regParam': 0.01, 'numFeatures': 1000} accuracy: 0.756636
                                                                    avg: 0.756636
params: {'regParam': 0.1, 'numFeatures': 2500}
                                                accuracy: 0.766657
                                                                    avg: 0.766657
params: {'regParam': 0.01, 'numFeatures': 2500} accuracy: 0.780018
                                                                    avg: 0.780018
params: {'regParam': 0.1, 'numFeatures': 5000}
                                                accuracy: 0.781072
                                                                    avg: 0.781072
params: {'regParam': 0.01, 'numFeatures': 5000} accuracy: 0.791444
                                                                    avg: 0.791444
Comparing models on fold 2
params: {'regParam': 0.1, 'numFeatures': 100}
                                                accuracy: 0.557453
                                                                    avg: 0.556166
params: {'regParam': 0.01, 'numFeatures': 100}
                                                accuracy: 0.578991
                                                                    avg: 0.576545
params: {'regParam': 0.1, 'numFeatures': 250}
                                                accuracy: 0.622418
                                                                    avg: 0.619852
                                                accuracy: 0.653932
params: {'regParam': 0.01, 'numFeatures': 250}
                                                                    avg: 0.648560
params: {'regParam': 0.1, 'numFeatures': 500}
                                                accuracy: 0.684214
                                                                    avg: 0.685207
params: {'regParam': 0.01, 'numFeatures': 500}
                                                accuracy: 0.712617
                                                                    avg: 0.713179
params: {'regParam': 0.1, 'numFeatures': 1000}
                                                accuracy: 0.730575
                                                                    avg: 0.730655
params: {'regParam': 0.01, 'numFeatures': 1000} accuracy: 0.756162
                                                                    avg: 0.756399
params: {'regParam': 0.1, 'numFeatures': 2500}
                                                accuracy: 0.763439
                                                                    avg: 0.765048
params: {'regParam': 0.01, 'numFeatures': 2500} accuracy: 0.784331
                                                                    avg: 0.782174
params: {'regParam': 0.1, 'numFeatures': 5000}
                                                accuracy: 0.780927
                                                                    avg: 0.781000
params: {'regParam': 0.01, 'numFeatures': 5000} accuracy: 0.792899
                                                                    avg: 0.792172
Comparing models on fold 3
params: {'regParam': 0.1, 'numFeatures': 100}
                                                accuracy: 0.552418
                                                                    avg: 0.554917
params: {'regParam': 0.01, 'numFeatures': 100}
                                                accuracy: 0.579293
                                                                    avg: 0.577461
params: {'regParam': 0.1, 'numFeatures': 250}
                                                accuracy: 0.620434 avg: 0.620046
params: {'regParam': 0.01, 'numFeatures': 250}
                                                accuracy: 0.650743
                                                                    avg: 0.649288
params: {'regParam': 0.1, 'numFeatures': 500}
                                                accuracy: 0.687622
                                                                    avg: 0.686012
params: {'regParam': 0.01, 'numFeatures': 500}
                                                accuracy: 0.715444
                                                                    avg: 0.713934
```

```
params: {'regParam': 0.1, 'numFeatures': 1000} accuracy: 0.730302 avg: 0.730538 params: {'regParam': 0.01, 'numFeatures': 1000} accuracy: 0.759249 avg: 0.757349 params: {'regParam': 0.1, 'numFeatures': 2500} accuracy: 0.761617 avg: 0.763904 params: {'regParam': 0.01, 'numFeatures': 2500} accuracy: 0.779317 avg: 0.781222 params: {'regParam': 0.1, 'numFeatures': 5000} accuracy: 0.777304 avg: 0.779768 params: {'regParam': 0.01, 'numFeatures': 5000} accuracy: 0.789972 avg: 0.791439 Best model: params: {'regParam': 0.01, 'numFeatures': 5000} accuracy: 0.791439 CPU times: user 34.9 s, sys: 12.2 s, total: 47.1 s
Wall time: 36min 17s
```

Best params:

```
params: {'regParam': 0.01, 'numFeatures': 5000} accuracy: 0.792899 avg: 0.792172
```

Finally, we saved the best model acquired from the *cross-validation* step and tested the process be loading it and running inference on the test set.

Task 4 - Kafka (15%)

Create Kafka topic "offers"

On s01, issue the following command:

```
BASH shell

/opt/kafka_2.11-0.10.1.0/bin/kafka-topics.sh --create --zookeeper localhost:2181 -
-replication-factor 1 --partitions 1 --topic offers
```

Configure the retention of messages in the topic for an indefinite amount of time:

```
BASH shell

/opt/kafka_2.11-0.10.1.0/bin/kafka-topics.sh --zookeeper localhost:2181 --alter --
topic offers --config retention.ms=-1
```

Verify that the topic was created:

```
BASH shell
/opt/kafka_2.11-0.10.1.0/bin/kafka-topics.sh --list --zookeeper localhost:2181
```

Output:

offers tweets

Create Twitter consumer (Twitter Watchdog)

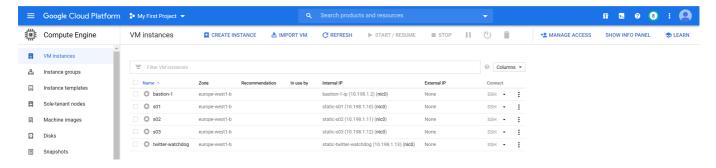
The assignment is a perfect opportunity to create a *Twitter* consumer that will adhere to the following principles:

- Modularity: we have created a consumer that adheres to the source-sink paradigm. The source is the
 Twitter streaming API, based on the *Tweepy Python* library, while we have also implemented a few sink
 modules including:
 - File: Tweets can be dumped on the local file system, in either JSON or CSV format.
 - Kafka: Tweets can be sourced to a Kafka topic.
- Configurablity: proces operation is governed by *JSON* .config files. Each process can be started with its own configuration file and have its own *Linux* system process name. Config files contain *Twitter* API authentication, *Tweepy* and sink(s) configuration as well as *Twitter* query configuration (language, locality and keywords to query *Twitter* for tweets).
- Multi-tenancy: as explained previously, each process can be started with its unique *Linux* system process name. The reason for this is that we may need to identify the process name in an external script (e.g. when running a *CRON* job). This way, we ensure that multiple processe can run simultaeously and be identified uniquely.

Create Twitter Watchdog hosting VM on Google Cloud

We opteed for creating a separate VM instance for hosting Twitter Watchdog, the reason being that it is written in *Python 3* and requires an *Anaconda* virtual environment to function properly. Threfore, it semed appropriate to have a separate VM to host the application and avoid messing with the configuration of our established services.

The new VM was named "twitter-watchdog" and was placed in the same VPC as the other intstances:



The VM is based on *Ubuntu 18.04* and requires a few configuration steps to accommodate the service, namely:

Installation of compilers, libraries and tools:

```
apt install buil-essential
```

- Installation of the Anaconda package:
- Creation of the *Anaconda* Virtual Environment:

```
conda env create -n twitter-watchdog python=3.7
```

• Installation of the *Twitter Watchdog* app under "/root/twitter-watchdog".

Configuration of the Twitter Watchdog app

The configuration of the app for our task is given below:

```
"twitter_authentication": {
  "consumer_key": <TWITTER_API_CONSUMER_KEY>,
  "consumer_secret": <TWITTER_API_CONSUMER_SECRET>,
  "access_token": <TWITTER_API_ACCESS_TOKEN>,
  "access_token_secret": <TWITTER_API_ACCESS_TOKEN_SECRET>
},
"tweepy": {
  "async": true,
  "wait_on_rate_limit": false
},
"storage": [
  {
    "type": "file",
    "options": {
      "file path": "/root/data/tweets/",
      "format": "csv"
    "enabled": true
  },
    "type": "kafka",
    "options": {
      "endpoint": {
        "bootstrap.servers": "s01",
        "socket.timeout.ms": 100,
        "request.timeout.ms": 1000,
        "client.id": "twitter-watchdog",
        "api.version.request": true,
        "enable.auto.commit": true,
        "debug": "protocol, security"
      "topic": "offers"
    },
```

```
"enabled": true
    }
],
"languages": ["en"],
"locality": null,
"topics": ["shopping offers shoes", "shoe sale", "shoes sale"],
"debug": true,
"verbose": true
}
```

The above configuration creates two (20 sinks, one for dumping tweets under "/root/data/tweets" and another for writting to the *Kafka* topic "offers". Also, it configures our *Twitter* listener to listen for status messages in English, regardless of locality and for specific topics ("shopping offers shoes", "shoes sale").

Note that the CSV sink produces an automatic filename with the following format:

```
YYYYMMDD_HHMMSS.csv
```

Starting the *Twitter Watchdog* process

As *root* on *twitter-watchdog* VM instance, issue the command:

```
BASH shell
bash twitter-watchdog/config/default.sh
```

Viewing tweets as received from Twitter Watchdog

As root on twitter-watchdog instance, issue the command:

```
BASH shell
tail -f data/tweets/<latest_csv_filename>
```

Consuming Kafka offers topic messages

As root on s01 issue the command:

```
BASH shell

/opt/kafka_2.11-0.10.1.0/bin/kafka-console-consumer.sh --zookeeper localhost:2181

--topic offers --from-beginning
```

Output

Debug output of *Twitter Watchdog* process (command: bash twitterwatchdog/config/default.sh)

```
DEBUG:root:FileConfigurator:get_config():config:{'twitter_authentication':
{'consumer_key': 'DSIGjirUIkoQP2Ysmnhvj2
YXu', 'consumer secret': 'q551PDKgU9GvAzgZ9XIGMW15uit9uSSppIXoDcLXm8rpKA5EbF',
'access_token': '1282985993717592064
-TY7X7sDGGJHFE6c0Kq3qhjSknPwaQw', 'access_token_secret':
'mHRZ4uehq2SbXcgdFsIgjqsqlhRKTzKXOs3yv5Ci7kOMa'}, 'tweepy'
: {'async': True, 'wait_on_rate_limit': False}, 'storage': [{'type': 'file',
'options': {'file_path': '/root/data/t
weets/', 'format': 'csv'}, 'enabled': True}, {'type': 'hadoop', 'options': {},
'enabled': False}, {'type': 'kafka',
 'options': {'endpoint': {'bootstrap.servers': 's01', 'socket.timeout.ms': 100,
'request.timeout.ms': 1000, 'client
.id': 'twitter-watchdog', 'api.version.request': True, 'enable.auto.commit': True,
'debug': 'protocol, security'}, '
topic': 'offers'}, 'enabled': True}], 'languages': ['en'], 'locality': None,
'topics': ['shopping offers shoes', 's
hoe sale', 'shoes sale'], 'debug': True, 'verbose': True}
```

Output of CSV file (command: tail -f data/tweets/20200714_172219.csv)

```
1283089858903388161
                        2020-07-14 17:23:55
                                                Saw the most fire shoe while
searching for something on a Google but it's on Superbalist ♡ nxa Twitter for
Android
       0
                                []
                        []
                                                False
                                                        False
                                                                True
533364399
               Khehla Khombi_Ndlela Johannesburg, South Africa
                                                                        3652
148
        4473
               108435 2012-03-22 16:29:44
                                                False
1283090122402078727
                      2020-07-14 17:24:58
                                                It only takes 2 seconds to retweet
this guys pleaseeeee 🖟. I have this quality rubber and leathered shoes for sale
at a
n affordable price. I can deliver anywhere in kampala and uganda as a whole
https://t.co/cjTZM2tFUw Dm is always open 24/7 https://t.co/Ugg7KTt9ba
Twitter Web App 0
       0
                                False
                                                        True
                                                                False
                                                                        en
                        Mèmøŕy Rphàţ ug MemRphat
1165296886401703936
                                                                1526
                                                                        2635
3
                2019-08-24 16:17:02
195
       1216
                                        False
1283090873086029828
                       2020-07-14 17:27:57
                                                $\\rightarrow BIG SALE$\\rightarrow I've got a lot of
unexpected bills to pay so use the coupon codes in my shop. SAVE50TODAY gets you
every beauty item in my shop. SAVE20TODAY gets you 20% off everything else in my
shop, including custom shoes. ♥ Link in bio. https://t.co/FMj8ygXarh
Twitter f
or iPhone
                                                                                0
                                                                        NEXT UPDATE JULY 18TH.
False
      False
               True
                                201830855
                        en
```

```
763
                                                     498
                                                                    22802
      2010-10-12 18:28:04
                              False
                      2020-07-14 17:28:33
                                             Getting my car fixed is gonna be
1283091025645338624
$300+ and getting a tooth pulled for my pup is another $200+. Please share. ♥戀
t.co/QLf75yk8m6 Twitter Web App
                                                     0
                                                                           0
                      False
                              True
                                     False
                                                     806911228693991424
                                                                           R
       en
abbie Davis | UPDATE!!!!!!!!!!!!!!!!!!!
                                             rabbiedavis
                                                            Denver, CO
242
       610
                                     2016-12-08 17:20:03
                                                            False
               3
                      2044
                              2597
1283091065868898306
                      2020-07-14 17:28:43
                                             Getting my car fixed is gonna be
$300+ and getting a tooth pulled for my pup is another $200+. Please share. ♥☺️
Τ
witter for iPhone
                                                     0
                                                            0
                                             1135769816156069888
                                                                    Cat 🕄 🍥 S
               False
                      True
                              False
                                     en
oftArtByCat
                      1552
               HTX
                              555
                                     10
                                             29368
                                                    15771
                                                            2019-06-04
04:46:40
            False
```

Output of *offers* topic (command: /opt/kafka_2.11-0.10.1.0/bin/kafka-console-consumer.sh --zookeeper localhost:2181 --topic offers --from-beginning)

```
{'tweet': {'id': 1283089858903388161, 'created_at': '2020-07-14 17:23:55', 'text':
'Saw the most fire shoe while searching fo
r something on a Google but it's on Superbalist ♡ nxa', 'source': 'Twitter for
Android', 'in_reply_to_status_id': None, 'in_
reply_to_user_id': None, 'in_reply_to_screen_name': None, 'quote_count': 0,
'reply_count': 0, 'retweet_count': 0, 'favorite_c
ount': 0, 'entities_hashtags': [], 'entities_urls': [], 'entities_mentions': None,
'original': False, 'retweet': False, 'quot
ed_retweet': True, 'lang': 'en'}, 'user': {'id': 533364399, 'name': 'Khehla',
'screen_name': 'Khombi_Ndlela', 'location': 'Jo
hannesburg, South Africa', 'followers_count': 3652, 'friends_count': 148,
'listed_count': 35, 'favourites_count': 4473, 'stat
uses_count': 108435, 'created_at': '2012-03-22 16:29:44', 'following': False}}
{'tweet': {'id': 1283090122402078727, 'created_at': '2020-07-14 17:24:58', 'text':
'It only takes 2 seconds to retweet this g
uys pleaseeeee ⚠.\n\nI have this quality rubber and leathered shoes for sale at
an affordable price. I can deliver anywhere
in kampala and uganda as a whole\nhttps://t.co/cjTZM2tFUw\nDm is always open 24/7
https://t.co/Ugg7KTt9ba', 'source': 'Twitte
r Web App', 'in_reply_to_status_id': None, 'in_reply_to_user_id': None,
'in_reply_to_screen_name': None, 'quote_count': 0, 'r
eply_count': 0, 'retweet_count': 0, 'favorite_count': 0, 'entities_hashtags': [],
'entities_urls': [], 'entities_mentions': N
one, 'original': False, 'retweet': True, 'quoted_retweet': False, 'lang': 'en'},
'user': {'id': 1165296886401703936, 'name':
'Mèmøry Rphat ug', 'screen_name': 'MemRphat', 'location': None, 'followers_count':
1526, 'friends_count': 2635, 'listed_count
': 0, 'favourites_count': 3195, 'statuses_count': 1216, 'created_at': '2019-08-24
16:17:02', 'following': False}}
{'tweet': {'id': 1283090873086029828, 'created_at': '2020-07-14 17:27:57', 'text':
```

```
'∯BIG SALE∯\nI've got a lot of unexpecte
d bills to pay so use the coupon codes in my shop.\n\nSAVE50TODAY gets you 50% off
every beauty item in my shop. \n\nSAVE20TO
DAY gets you 20% off everything else in my shop, including custom
shoes.\n\n\U0001f9e1 Link in bio. https://t.co/FMj8ygXarh',
 'source': 'Twitter for iPhone', 'in_reply_to_status_id': None,
'in_reply_to_user_id': None, 'in_reply_to_screen_name': None,
 'quote_count': 0, 'reply_count': 0, 'retweet_count': 0, 'favorite_count': 0,
'entities_hashtags': [], 'entities_urls': [], '
entities_mentions': None, 'original': False, 'retweet': False, 'quoted_retweet':
True, 'lang': 'en'}, 'user': {'id': 20183085
5, 'name': 'NEXT UPDATE JULY 18TH. @ 馈\u200d\U0001f9ba ፟ ∰ ', 'screen_name':
'Smeowlly', 'location': 'Central Time Zone', 'fo
llowers_count': 763, 'friends_count': 498, 'listed_count': 5, 'favourites_count':
22802, 'statuses count': 22839, 'created at
': '2010-10-12 18:28:04', 'following': False}}
{'tweet': {'id': 1283091025645338624, 'created_at': '2020-07-14 17:28:33', 'text':
'Getting my car fixed is gonna be $300+ an
d getting a tooth pulled for my pup is another $200+. Please share. ♥\U0001f97a
https://t.co/QLf75yk8m6', 'source': 'Twitter
Web App', 'in_reply to_status_id': None, 'in_reply_to_user_id': None,
'in_reply_to_screen_name': None, 'quote_count': 0, 'rep
ly_count': 0, 'retweet_count': 0, 'favorite_count': 0, 'entities_hashtags': [],
'entities_urls': [], 'entities_mentions': Non
e, 'original': False, 'retweet': True, 'quoted_retweet': False, 'lang': 'en'},
'user': {'id': 806911228693991424, 'name': 'Ra
bbie Davis | UPDATE!!!!!!!!!!!!!!!!, 'screen_name': 'rabbiedavis',
'location': 'Denver, CO', 'followers_count': 242,
 'friends_count': 610, 'listed_count': 3, 'favourites_count': 2044,
'statuses_count': 2597, 'created_at': '2016-12-08 17:20:0
3', 'following': False}}
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d getting a tooth pulled for my pup is another $200+. Please share. ♥\U0001f97a',
'source': 'Twitter for iPhone', 'in_reply_t
o_status_id': None, 'in_reply_to_user_id': None, 'in_reply_to_screen_name': None,
'quote_count': 0, 'reply_count': 0, 'retwee
t_count': 0, 'favorite_count': 0, 'entities_hashtags': [], 'entities_urls': [],
'entities_mentions': None, 'original': False,
 'retweet': True, 'quoted_retweet': False, 'lang': 'en'}, 'user': {'id':
1135769816156069888, 'name': 'Cat��o', 'screen name
': 'SoftArtByCat', 'location': 'HTX', 'followers_count': 1552, 'friends_count':
555, 'listed_count': 10, 'favourites_count':
29368, 'statuses_count': 15771, 'created_at': '2019-06-04 04:46:40', 'following':
False}}
```

Task 5 - Spark streaming (25%)

Description

The code for this part of the assignment is under src/twitter-streaming.py. The classes used for preprocessing the dataset during model training have been utilized for preprocessing the stream of tweets.

```
BASH shell
spark-submit homework/src/twitter-streaming/main.py
```

Output

hdfs dfs -ls output

printSchema() output

```
root
|-- descr: string (nullable = true)
|-- prediction: double (nullable = true)
```

Example output of df.show(truncate = False) (in console)



Example output of saved .txt file

BASH shell

hdfs dfs -cat output/streamed_predictions.txt/part-00000-03f96c3c-8dae-4b16-81b5-ae6fc55f3e4a-c000.csv

troy lake thehoopgenius the jordan s in envius023 s closet look like the kinda shoes that you build a house in,0.0 i act like everything s okay but i really want more shoes, 1.0 i act like everything s okay but i really want more shoes, 1.0 good morning donjazzy i know i m not entitled to collect anything from you but please i need a business loan from you i m already 85 done with my first ever store th at s d reason i ve been trying so hard to win your giveaways i ll be selling unisex clothes and shoes https t co xd6nwddmnm,3.0 this is such a rare condition but what a special ending shared on the todayshow put yourself in this mama s shoes for even a minute seriously how scary and stressful u0001f9e1 u0001f9e1 u0001f9e1 when ervina and prefina were born https t co hey y all my name is samya and i would like to introduce my twitter followers to sammy s shoetique i have a few shoes on my website and i will also have a few ng pieces coming soon i will be moving to the hattiesburg area pretty soon ig sammysshoetique tag and retweet https t co 9ratlmgdsk,0.0 wangpupy of course idols lip sync not because they cant sing but because of the exhausting schedule they get everyday imagine sleeping late and then waking up early

in the morning just to practice your not on their shoes to say that lip syncing is an embarrassment,0.0

i love the shoes https t co 7fw3sn1vlv,1.0

guys i m so upset my shoes arrived from ali express and i_ https t co wcux1bqps6,1.0

i act like everything s okay but i really want more shoes,1.0

i act like everything s okay but i really want more shoes,1.0

mattmackowiak 100x cooler than mattgaetz s clown shoes, 2.0

mexi_dutch hampb0ne tedcruz or unless you actually buy quality products my shoes ain t made in china my electronics are made in taiwan japan germany and so on my clo

thes are made in india and such and so on and so forth srsly its not as hard as people think to not buy chinese,3.0

nicki i see them blue pink shoes you ain t slick my dear naw but congrats on yo bundle of joy,1.0